

Dataset :

Super Store Sales analysis:

Load Data Set :

Super Store Sales analysis																		
	Row ID	Order ID	Order Date	Ship Date	Ship Mode	Customer ID	Customer Name	Segment	Country	City	...	Postal Code	Region	Product ID	Category	Sub-Category	Product Name	
0	1	CA-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderson	...	42420	South	FUR-BO-10001798	Furniture	Bookcases	Business & Industrial: Bookcases & Storage	
1	2	CA-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderson	...	42420	South	FUR-CH-10000454	Furniture	Chairs	Hon Deluxe Fabricated Steel Stackable Chairs	
2	3	CA-138688	2016-06-12	2016-06-16	Second Class	DV-13045	Darrin Van Huff	Corporate	United States	Los Angeles	...	90036	West	OFF-LA-10000240	Office Supplies	Labels	Self Adhesive Address Labels for Typewriter	
3	4	US-108966	2015-10-11	2015-10-18	Standard Class	SO-20335	Sean O'Donnell	Consumer	United States	Fort Lauderdale	...	33311	South	FUR-TA-10000577	Furniture	Tables	Bretford CR450 Series Slim Rectangular	

Perform Task :

1. Shape

```
df.shape # how much row and column  
(9994, 21)
```

2. Columns

```
df.columns # Columns name  
Index(['Row ID', 'Order ID', 'Order Date', 'Ship Date', 'Ship Mode',  
       'Customer ID', 'Customer Name', 'Segment', 'Country', 'City', 'State',  
       'Postal Code', 'Region', 'Product ID', 'Category', 'Sub-Category',  
       'Product Name', 'Sales', 'Quantity', 'Discount', 'Profit'],  
      dtype='object')
```

3. Information of Data

```
df.info() # table information
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 9994 entries, 0 to 9993
Data columns (total 21 columns):
 #   Column          Non-Null Count  Dtype  
--- 
 0   Row ID          9994 non-null    int64  
 1   Order ID        9994 non-null    object  
 2   Order Date      9994 non-null    object  
 3   Ship Date       9994 non-null    object  
 4   Ship Mode       9994 non-null    object  
 5   Customer ID     9994 non-null    object  
 6   Customer Name   9994 non-null    object  
 7   Segment          9994 non-null    object  
 8   Country          9994 non-null    object  
 9   City              9994 non-null    object  
 10  State             9994 non-null    object  
 11  Postal Code     9994 non-null    int64  
 12  Region            9994 non-null    object  
 13  Product ID      9994 non-null    object  
 14  Category          9994 non-null    object  
 15  Sub-Category     9994 non-null    object  
 16  Product Name     9994 non-null    object  
 17  Sales             9994 non-null    float64 
 18  Quantity          9994 non-null    int64  
 19  Discount          9994 non-null    float64 
 20  Profit             9994 non-null    float64 
dtypes: float64(3), int64(3), object(15)
memory usage: 1.6+ MB
```

5. Business Performance

```
# Total sales , profit, orders , unique customers
total_sales = df['Sales'].sum()
total_profits = df['Profit'].sum()
unique_customers = df['Customer Name'].nunique()
total_orders = df['Order ID'].nunique()
df.head(793)

total_sales , total_profits , unique_customers, total_orders
(np.float64(2297200.860300003), np.float64(286397.0217), 793, 5009)
```

4. Duplicate value

```
df.duplicated().sum() # Find duplicate in given data
np.int64(0)
```

7. Minimum Sales

```
# find minimum sales
df.groupby('Category')['Sales'].min()

Category
Furniture      1.892
Office Supplies 0.444
Technology      0.990
Name: Sales, dtype: float64
```

6. Maximum Sales

```
# find maximum sales
df.groupby('Category')['Sales'].max()

Category
Furniture      4416.174
Office Supplies 9892.740
Technology      22638.480
Name: Sales, dtype: float64
```

8. Total sales by Year and Month

```
# aggregates total sales by Year and Month
df['Year'] = df['Order Date'].dt.year
df['Month'] = df['Order Date'].dt.month_name()
df.groupby('Year')['Sales'].sum()
df.groupby('Month')['Sales'].sum().sort_values(ascending = False)

Month
November      352461.0710
December       325293.5035
September      307649.9457
March          205005.4888
October         200322.9847
August          159044.0630
May             155028.8117
June             152718.6793
July             147238.0970
April            137762.1286
January          94924.8356
February         59751.2514
Name: Sales, dtype: float64
```

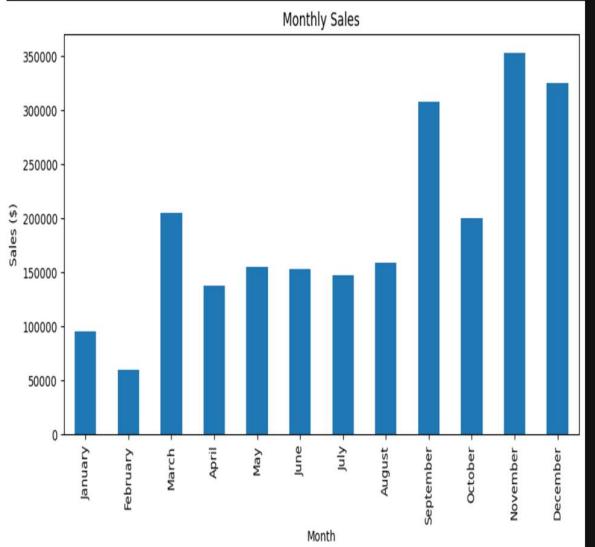
9. Finds the mean profit

```
# finds the mean profit
df.groupby('Region')['Profit'].mean()

Region
Central      17.092709
East          32.135808
South          28.857673
West          33.849032
Name: Profit, dtype: float64
```

10. Monthly sales using plot

```
# monthly sales using plot
monthly_sales = df.groupby('Month')['Sales'].sum()
monthly_sales = monthly_sales.reindex([
    'January', 'February', 'March', 'April', 'May', 'June',
    'July', 'August', 'September', 'October', 'November', 'December'
])
monthly_sales.plot(kind='bar', figsize=(10,5), title='Monthly Sales')
plt.ylabel('Sales ($)')
plt.show()
```



11. Generate highest profit

```
# generate the highest overall profit
df.groupby('Sub-Category')['Profit'].sum().sort_values(ascending = False)
```

Sub-Category	Profit
Copiers	55617.8249
Phones	44515.7396
Accessories	41936.6357
Paper	34053.5693
Binders	38221.7633
Chairs	26590.1663
Storage	21228.8264
Appliances	18138.0864
Furnishings	19959.1436
Envelopes	6964.1767
Art	6527.7870
Labels	5546.2540
Machines	3384.7569
Fasteners	949.5182
Supplies	-1189.0995
Bookcases	-3472.5566
Tables	-17725.4811

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Labels	5546.2540
Machines	3384.7569
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Supplies	-1189.0995
Bookcases	-3472.5560
Tables	-1725.4811

Name: Profit, dtype: float64

12. use for date like [year – month- day]

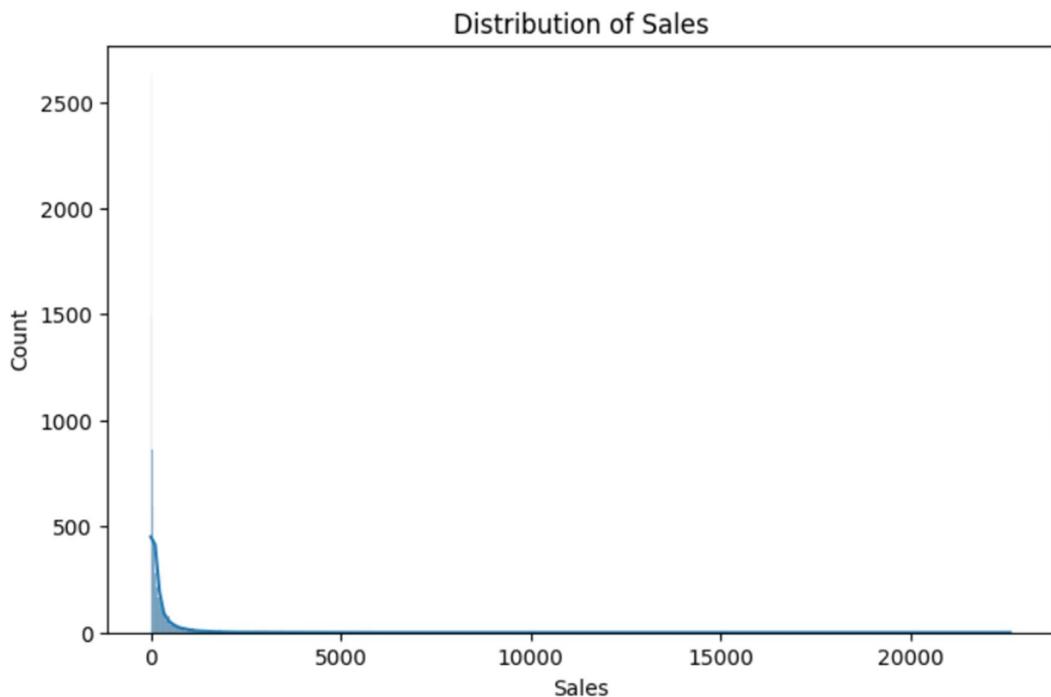
```
# Use for date Like [ year-month-day]
df['Order Date'] = pd.to_datetime(df['Order Date'])
df['Ship Date'] = pd.to_datetime(df['Ship Date'])
df.head(2) # Use for date like [ year-month-day]
```

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1	2	CA-152156	2016-11-08	2016-11-11	Second Class	CG-12520	Claire Gute	Consumer	United States	Henderson	...	42420	South	FUR-CH-10000454	Furniture

2 rows × 21 columns

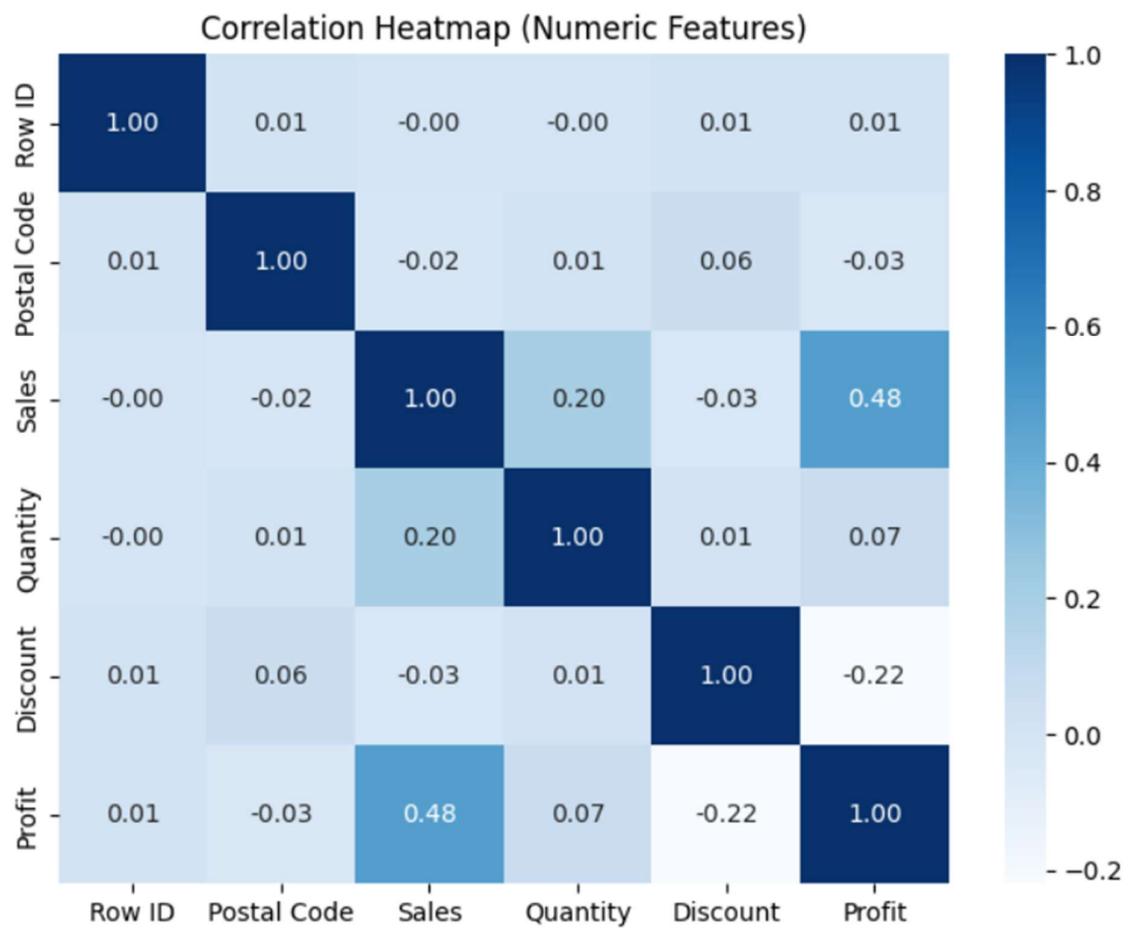
13. Distribution of Sales

```
plt.figure(figsize=(8,5))
sns.histplot(df['Sales'], kde=True)
plt.title("Distribution of Sales")
plt.show()
```



14. Correlation Heatmap(Numeric Features)

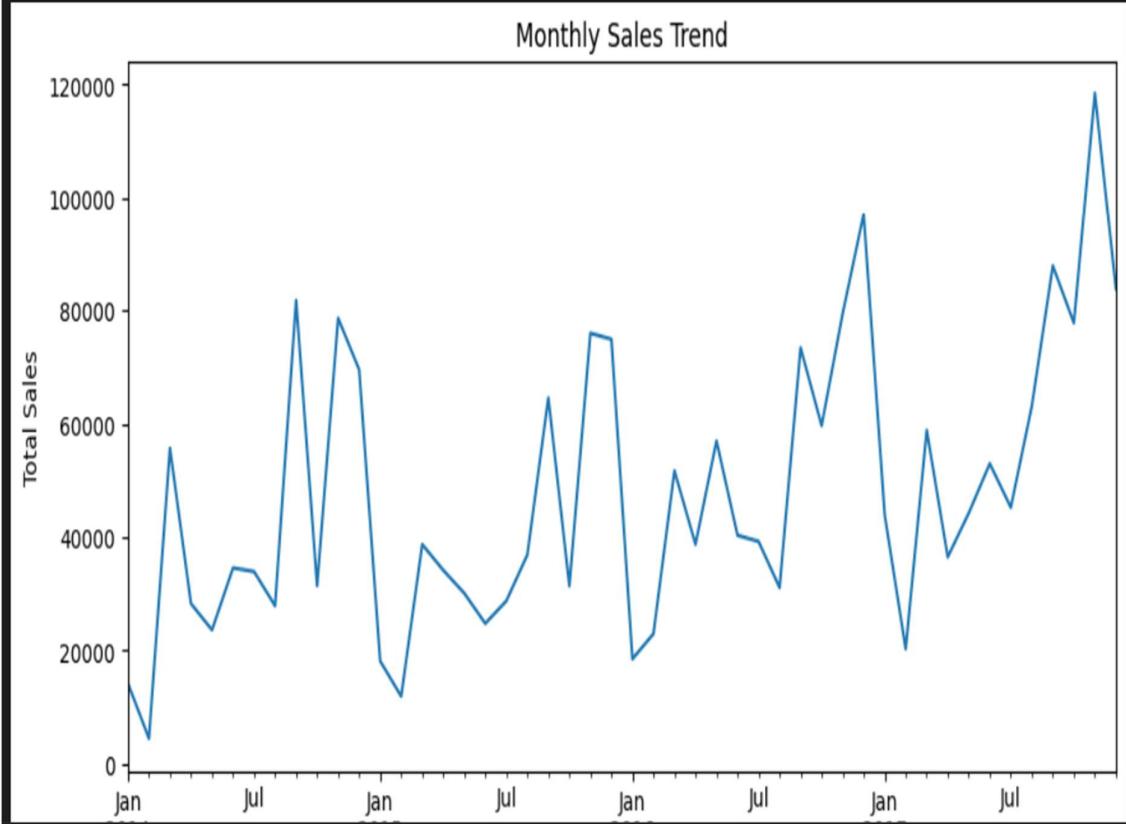
```
plt.figure(figsize=(8,6))
numeric_df = df.select_dtypes(include=['int64', 'float64'])
sns.heatmap(numeric_df.corr(), annot=True, fmt=".2f", cmap="Blues")
plt.title("Correlation Heatmap (Numeric Features)")
plt.show()
```



15. Monthly Sales Trend

```
df_time = df.groupby(df["Order Date"].dt.to_period("M"))["Sales"].sum()

plt.figure(figsize=(10,5))
df_time.plot()
plt.title("Monthly Sales Trend")
plt.xlabel("Year-Month")
plt.ylabel("Total Sales")
plt.show()
```



16. Compare Sales Vs Profit

```
plt.figure(figsize=(7,5))
sns.scatterplot(x="Sales", y="Profit", data=df)
plt.title("Sales vs Profit")
plt.show()
```



17. Total Profit by Segment

```
plt.figure(figsize=(7,5))
sns.barplot(x="Segment", y="Profit", data=df, estimator=sum, palette="viridis")
plt.title("Total Profit by Segment")
plt.show()

c:\users\tarpi\appdata\local\temp\ipykernel_1964\105888946.py:2: FutureWarning:
Passing 'palette' without assigning 'hue' is deprecated and will be removed in v0.14.0. Assign the 'x' variable to 'hue' and set 'legend=False' for the same effect.

sns.barplot(x="Segment", y="Profit", data=df, estimator=sum, palette="viridis")
```

